

AC215 - Team Dashers



### **Background**

### **01. Problem Statement**

**02. Target Audience** 

- Can't decide what to eat?
- Limited time to cook?
- Tired of spoiled food in your fridge?

You need a personal meal planning assistant! Take a picture of your ingredients, you will get a **streamline meal planning** with:



Recommended recipe with estimated cooking time and calories



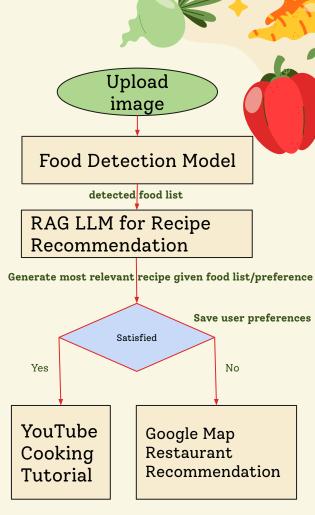
Route suggestions for dining out

Food expiry tracker









## **Unique Value Proposition**

# Smart Integration of AI Technologies

# Competitors

### **Our Benefits**

#### **Object Detection:**

update inventory effortlessly

#### LLM:

personalized recipe recommendation

#### API calling agent:

Dine-out: Google map

Cook: YouTube

#### Fail to offer end-to-end solution:

- No ingredient expiry tracker
- No "cook-or-dine" comparison
- No personalized food profile
- No ingredient-driven recipe recommendation

#### Efficiency:

- Automate ingredient tracking
- Cost, Time, Health

#### **Guidance:**

- Video cooking tutorials
- Map navigation to restaurants

#### **Engaging:**

- Decision-Making Aid
- Weekly Summary Visuals







- Model Selection and Performance Optimization
  - Core Features Implemented
    - **■** Food Detection + Recipe Generation
  - Food Detection: GPT-4o-mini
    - Contextual Understanding
      - educated guess on partially visible/ambiguous items
    - Textual Descriptions
      - generate descriptive text
    - Flexibility and Accuracy



- **Model Selection and Performance Optimization** 
  - Object Detection Model Selection

**GPT** 

Input can you list the name of food from the picture? From the picture of the refrigerator, here's a list of foods I can identify: 1. Red hell penners 2. Apples (red) 3. Strawberries 4. Lemons 5. Eggs (in a carton) 6 Corn on the coh 7. Pumpkin or butternut squash slices 8. Oranges 10. Grapes (green) 11 Watermalon (email 13. Napa cabbage 14. Green bell pepper 15. Cucumbers 16. Pickles or jarred vegetables (in the bot! J. drawer)

EfficientDet-LiteO



YOLOV8s (Finetuned)

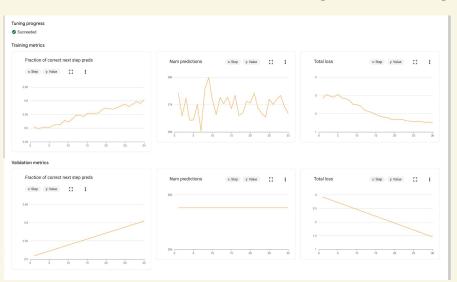


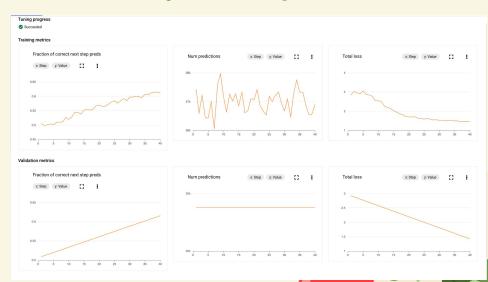


- Recipe Generation: RAG with Finetuned Gemini 1.5 Flash
  - Fine Tuning Motivation
    - Improve Contextual Understanding/Refine Recipe-Specific Embeddings
    - Prevent Hallucinations
  - Training Data
    - Raw Data: Recipes data with titles, ingredients, and preparation directions
    - Training Data: 5000 question-answer pairs
      - Q: "How do you prepare [recipe title] using these ingredients: [ingredients]?"
      - A: corresponding recipe directions.



- Recipe Generation: RAG with Finetuned Gemini 1.5 Flash
  - o Fine Tuning Process
    - Base Model: gemini-1.5-flash-002
    - Hyperparameters
      - v1: epochs=3, adapter\_size=4, learning\_rate\_multiplier=1.0
      - v2: epochs=4, adapter\_size=4, learning\_rate\_multiplier=0.9







#### RAG

• Recipe vector database (example in the table)

Title	Ingredients	Unquantified ingredients (Key)	Directions (Value)	Style
Jewell Ball's Chicken	["1 small jar chipped beef, cut up", "4 boned]	["beef", "chicken breasts", "cream of mushroom]	["Place chipped beef on bottom of baking dish]	Casserole

#### Self-RAG

- Auto-determine when to use vector embedding searching
  - Filter out irrelevant information when embedding queries
  - Extract an unquantified food list and styles (Query):ex. [dough, pizza source,...]





\*

### CLI example response

Enter your ingredients (or press Enter to use the default [broccoli, chicken, cheese]): [beef, corn]

### LLM Response:

Based on the recipes you provided, we can make a \*\*Corn Beef Casserole\*\*.

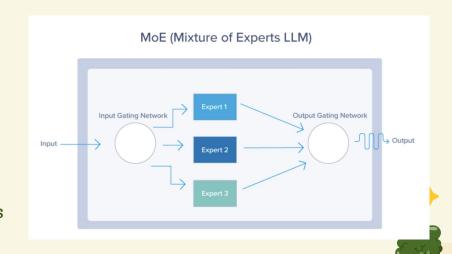
Here's what we'll need:

- \* \*\*Beef\*\* (from your ingredients)
- \* \*\*Corn\*\* (from your ingredients)
- \* \*\*1 (16 1/2 oz.) can whole kernel corn, drained\*\*
- \* \*\*1 (16 1/2 oz.) can cream-style corn\*\*
- \* \*\*1 (8 oz.) sour cream\*\*
- \* \*\*1 (8 1/2 oz.) pkg. Jiffy corn bread mix\*\*
- \* \*\*1 stick margarine\*\*
- \*\*Instructions:\*\*
- 1. \*\*Brown the beef\*\* in a skillet.
- 2. \*\*In a bowl, mix the beef, corns, sour cream, corn bread mix, and melted margarine.\*\*
- 3. \*\*Pour into a greased 8 x 8 x 2-inch pan.\*\*



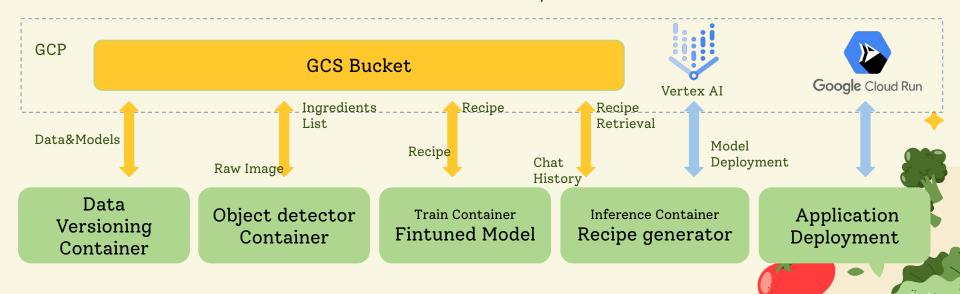
### **Technical Scalability**

- Gemini 1.5
  - Mixture-of-Experts (MoE)
     Architecture
    - Divides into smaller/ specialized networks
    - Activate a subset of experts for each input
- Serverless Computing: Google Cloud Run
  - Supports up to 80 concurrent requests (configurable to 1000 max)
  - Scalability tested with Google Kubernetes Engine





- GCP
  - Scalable cloud infrastructure; ideal for large data & model demands.
- Vertex AI
  - Streamlined model training/deployment/ monitoring
- Docker
  - Consistent environments across development/ production
- DVC
  - o Facilitates version control for datasets/ model files





### **Future Development**

1. Build Vertex AI pipeline

Automate the workflow

2. UI integration with WhatsApp

Easy to start for users

Bigger Picture?

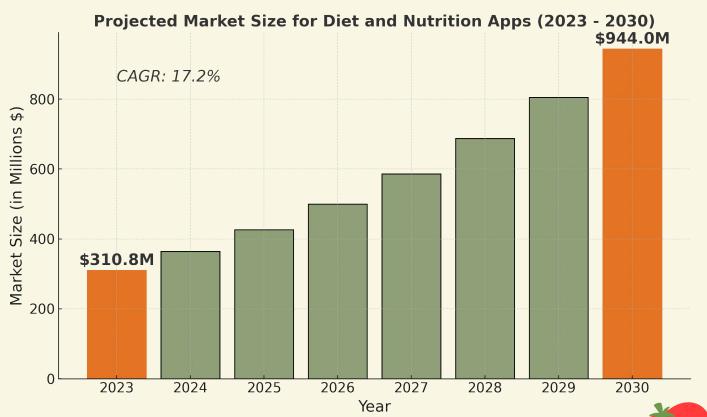








### **Market Growth**





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**Key drivers** 

Health Consciousness

Prevalence of Obesity

Lifestyle influenced chronic disorders:

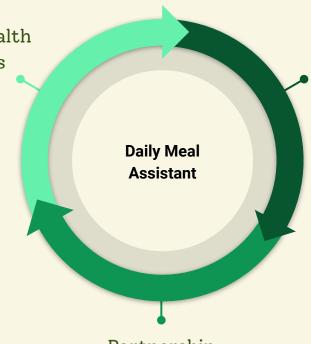
422 million individuals have diabetes globally in 2023

World Obesity Federation: 1 billion people have been estimated to be living with obesity by 2030 globally



Data Tracking for Health and Diet Outcomes

- 1. Allow user to set dietary goals related to weight management, diabetes control...
- 2. Collect user feedback for continuous improvements



Partnership with Food and Health Brands Develop Health-Focused Recipe Recommendations

- 1. Create Recipe Categories Targeting Chronic Conditions
- 2. Add Filters for Dietary and Health Goals

high-fiber

nutrient-dense

low-sugar





# Thank you!



Let's Q&A...



